

## Problem Set 2

Problems marked **(T)** are for discussions in Tutorial sessions.

1. **(T)** Are the matrices  $\begin{bmatrix} 1 & 2 \\ 4 & 8 \end{bmatrix}$ ,  $\begin{bmatrix} 0 & 0 \\ 1 & 2 \end{bmatrix}$  and  $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$  row-equivalent?

**Solution:**  $\begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ -4 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 4 & 8 \end{bmatrix}$ . So,  $\begin{bmatrix} 0 & 0 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ -4 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 4 & 8 \end{bmatrix}$ . The third matrix  $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$  is NOT row equivalent to either as column transformation is not allowed (or RREF are not the same).

2. Supply two examples each and explain their geometrical meaning.

- (a) Two linear equations in two variables with exactly one solution.

**Solution:**

$\left. \begin{array}{l} x + y = 2 \\ x - y = 0 \end{array} \right\}$ . They represent two lines in  $\mathbb{R}^2$  intersecting at a point.

- (b) Two linear equations in two variables with infinitely many solutions.

**Solution:**

$\left. \begin{array}{l} x + y = 2 \\ 2x + 2y = 4 \end{array} \right\}$ . They represent the same line in  $\mathbb{R}^2$ .

- (c) Two linear equations in two variables with no solutions.

**Solution:**

$\left. \begin{array}{l} x + y = 2 \\ 2x + 2y = 1 \end{array} \right\}$ . They represent two parallel lines in  $\mathbb{R}^2$ , no intersection.

- (d) Three linear equations in two variables with exactly one solution.

**Solution:**

$\left. \begin{array}{l} x + y = 2 \\ x - y = 0 \\ 2x - y = 1 \end{array} \right\}$ . They represent three lines in  $\mathbb{R}^2$  with a single point in common.

- (e) Three linear equations in two variables with no solutions.

**Solution:**

$\left. \begin{array}{l} x + y = 2 \\ x - y = 0 \\ 2x + 2y = 1 \end{array} \right\}$ . They represent three lines in  $\mathbb{R}^2$  with no point in common.

3. Suppose that  $\mathbf{x}$  and  $\mathbf{y}$  are two distinct solutions of the system  $A\mathbf{x} = \mathbf{b}$ . Prove that there are infinitely many solutions to this system, by showing that  $\lambda\mathbf{x} + (1 - \lambda)\mathbf{y}$  is also a solution, for each  $\lambda \in \mathbb{R}$ . Do you have a geometric interpretation?

**Solution:** If  $A$  is  $m \times n$  matrix then the line segment joining the two points in  $\mathbb{R}^n$  is also a solution. So, the system  $A\mathbf{x} = \mathbf{b}$  cannot have finite number of solutions over  $\mathbb{R}$ .

4. Let  $B$  be a square invertible matrix. Then, prove that the system  $A\mathbf{x} = \mathbf{b}$  and  $BA\mathbf{x} = B\mathbf{b}$  are row-equivalent.

**Solution:**

As  $B$  is invertible, there exists elementary matrices  $E_i$ 's such that  $B = E_1 E_2 \cdots E_k$ . Thus, the system  $BA\mathbf{x} = B\mathbf{b}$  is obtained from  $A\mathbf{x} = \mathbf{b}$  by  $k$  elementary row operations.

Conversely, as  $B^{-1} = E_k^{-1} E_{k-1}^{-1} \cdots E_1^{-1}$  and inverse of elementary matrices are also elementary matrices, we do obtain  $A\mathbf{x} = \mathbf{b}$  from  $BA\mathbf{x} = B\mathbf{b}$  by  $k$  elementary row operations. Thus, the above two systems are row equivalent.

5. [T] Suppose  $A\mathbf{x} = \mathbf{b}$  and  $C\mathbf{x} = \mathbf{b}$  have same solutions for every  $\mathbf{b}$ . Is it true that  $A = C$ ?

**Solution:** First, as  $A\mathbf{x} = \mathbf{b}$  and  $C\mathbf{x} = \mathbf{b}$  have same solutions,  $A$  and  $C$  have same shapes, that is, same number of rows and columns, as well as same null space,  $\text{Null}(A) = \{\mathbf{x} \in \mathbb{R}^n : A\mathbf{x} = \mathbf{0}\}$  (follows from taking  $\mathbf{b} = \mathbf{0}$ ).

Now, if we take  $\mathbf{b}$  to be the first column of  $A$  then  $\mathbf{x} = [1 \ 0 \ \dots \ 0]^T$  solves  $A\mathbf{x} = \mathbf{b}$  and therefore also solves  $C\mathbf{x} = \mathbf{b}$  which in turn implies that the first columns of  $A$  and  $C$  are same. Same argument holds for other columns as well. Thus  $A = C$ .

6. [T] Find matrices  $A$  and  $B$  with the given property or explain why you can not find them?

(a) The only solution to  $A\mathbf{x} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$  is  $\mathbf{x} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ .

**Solution:** Any  $A$  satisfying the given equation has to be a  $3 \times 2$  matrix. The linear system has a unique solution when rank of  $A$  is 2 and  $[1 \ 2 \ 3]^T \in \{A\mathbf{x} : \mathbf{x} \in \mathbb{R}^2\}$ . Among many possibilities, one such  $A$  is

$$\begin{bmatrix} 1 & 1 \\ 0 & 2 \\ 0 & 3 \end{bmatrix}.$$

(b) The only solution to  $B\mathbf{x} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$  is  $\mathbf{x} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$ .

**Solution:** Any  $B$  satisfying the given equation has to be a  $2 \times 3$  matrix which implies that the null space of  $B$ ,  $N(B)$ , can not be trivial and hence we either have an infinitely many solutions (when  $[0 \ 1]^T$  lies in the column space of  $B$ ) or no solution. Thus, finding a  $B$  that exhibits a unique solution is not possible.

7. Using Gauss Jordan method, find the inverse of  $\begin{bmatrix} 1 & 2 & 2 \\ 2 & 1 & 2 \\ 2 & 2 & 1 \end{bmatrix}$ .

**Solution:**

$$\left[ \begin{array}{ccc|ccc} 1 & 2 & 2 & 1 & 0 & 0 \\ 2 & 1 & 2 & 0 & 1 & 0 \\ 2 & 2 & 1 & 0 & 0 & 1 \end{array} \right] \xrightarrow[R_3 \leftarrow R_3 - 2R_1]{R_2 \leftarrow R_2 - 2R_1} \left[ \begin{array}{ccc|ccc} 1 & 2 & 2 & 1 & 0 & 0 \\ 0 & -3 & -2 & -2 & 1 & 0 \\ 0 & -2 & -3 & -2 & 0 & 1 \end{array} \right] \xrightarrow{R_3 \leftarrow R_3 - (2/3)R_2}$$

$$\begin{aligned}
& \left[ \begin{array}{ccc|ccc} 1 & 2 & 2 & 1 & 0 & 0 \\ 0 & -3 & -2 & -2 & 1 & 0 \\ 0 & 0 & -5/3 & -2/3 & -2/3 & 1 \end{array} \right] \xrightarrow{R_3 \leftarrow (-3/5)R_3} \left[ \begin{array}{ccc|ccc} 1 & 2 & 2 & 1 & 0 & 0 \\ 0 & -3 & -2 & -2 & 1 & 0 \\ 0 & 0 & 1 & 2/5 & 2/5 & -3/5 \end{array} \right] \\
& \xrightarrow[\begin{smallmatrix} R_2 \leftarrow R_2 + 2R_3 \\ R_1 \leftarrow R_1 - 2R_3 \end{smallmatrix}]{\begin{smallmatrix} R_2 \leftarrow R_2 + 2R_3 \\ R_1 \leftarrow R_1 - 2R_3 \end{smallmatrix}} \left[ \begin{array}{ccc|ccc} 1 & 2 & 0 & 1/5 & -4/5 & 6/5 \\ 0 & -3 & 0 & -6/5 & 9/5 & -6/5 \\ 0 & 0 & 1 & 2/5 & 2/5 & -3/5 \end{array} \right] \xrightarrow{R_2 \leftarrow (-1/3)R_2} \\
& \left[ \begin{array}{ccc|ccc} 1 & 2 & 0 & 1/5 & -4/5 & 6/5 \\ 0 & 1 & 0 & 2/5 & -3/5 & 2/5 \\ 0 & 0 & 1 & 2/5 & 2/5 & -3/5 \end{array} \right] \xrightarrow{R_1 \leftarrow R_1 - 2R_2} \left[ \begin{array}{ccc|ccc} 1 & 0 & 0 & -3/5 & 2/5 & 2/5 \\ 0 & 1 & 0 & 2/5 & -3/5 & 2/5 \\ 0 & 0 & 1 & 2/5 & 2/5 & -3/5 \end{array} \right]
\end{aligned}$$

Thus, the inverse is  $\begin{bmatrix} -3/5 & 2/5 & 2/5 \\ 2/5 & -3/5 & 2/5 \\ 2/5 & 2/5 & -3/5 \end{bmatrix}$ .

8. **(T)** Let  $B \in \mathbb{M}_n(\mathbb{R})$  be a real skew-symmetric matrix. Show that  $I - B$  is non singular.

**Solution:**

Let if possible  $I - B$  be singular. Then, the system  $(I - B)\mathbf{x} = \mathbf{0}$  has a non-trivial solution, say  $\mathbf{x}_0 \neq \mathbf{0}$ . Hence,  $B\mathbf{x}_0 = \mathbf{x}_0$ . Also,  $\mathbf{x}_0^T B\mathbf{x}_0 \in \mathbb{R}$  and hence

$$\mathbf{x}_0^T B\mathbf{x}_0 = (\mathbf{x}_0^T B\mathbf{x}_0)^T = \mathbf{x}_0^T B^T \mathbf{x}_0 = -\mathbf{x}_0^T B\mathbf{x}_0.$$

Thus,  $\mathbf{x}_0^T B\mathbf{x}_0 = 0$ . But,  $0 = \mathbf{x}_0^T B\mathbf{x}_0 = \mathbf{x}_0^T (B\mathbf{x}_0) = \mathbf{x}_0^T \mathbf{x}_0$  and hence  $\mathbf{x}_0 = \mathbf{0}$ .

9. For two  $n \times n$  matrices  $A$  and  $B$ , show that  $\det(AB) = \det(A)\det(B)$ .

**Solution:** First, suppose that  $A$  is singular. Then,  $AB$  is singular as well. We therefore have,  $\det(A) = 0$  and  $\det(AB) = 0$  which leads to  $\det(AB) = 0 = \det(A)\det(B)$ .

Now we assume that  $A$  is non-singular. Recall that, for a non-singular square matrix, the reduced row echelon form is the identity matrix,  $I$ . In other words, there exist elementary matrices  $E_s, E_{s-1}, \dots, E_2, E_1$  such that

$$A = E_s E_{s-1} \dots E_2 E_1 I.$$

We therefore have

$$AB = E_s E_{s-1} \dots E_2 E_1 B.$$

Thus the problem of showing  $\det(AB) = \det(A)\det(B)$  reduces to showing

- (a)  $\det(E_{ij}B) = \det(E_{ij})\det(B)$ .
- (b)  $\det(E_i(c)B) = \det(E_i(c))\det(B)$ .
- (c)  $\det(E_{ij}(c)B) = \det(E_{ij}(c))\det(B)$ .

Now, for the proof, we use the defining properties 1, 2 and 3 discussed in class. We have

- (a)  $\det(E_{ij}B) = -\det(B)$  (property 2)  $= \det(E_{ij})\det B$  (using  $\det(E_{ij}) = -\det(I) = -1$ , follows from properties 2 and 1).

- (b)  $\det(E_i(c)B) = c \det(B)$  (property 3a)  $= \det(E_i(c))\det(B)$  (using  $\det(E_i(c)) = c \det(I) = c$ , follows from properties 3a and 1).
- (c)  $\det(E_{ij}(c)B) = \det(B)$  (property 6)  $= \det(E_{ij}(c))\det(B)$  (using  $\det(E_{ij}(c)) = 1$ , follows from property 6 on the identity matrix).

The result now follows

$$\begin{aligned}
 \det(AB) &= \det(E_s E_{s-1} \dots E_2 E_1 B) = \det(E_s) \det(E_{s-1} \dots E_2 E_1 B) = \dots \\
 &= \det(E_s) \det(E_{s-1}) \dots \det(E_2) \det(E_1 B) \\
 &= \det(E_s) \det(E_{s-1}) \dots \det(E_2) \det(E_1) \det(B) \\
 &= \det(E_s) \det(E_{s-1}) \dots \det(E_2 E_1) \det(B) = \dots \\
 &= \det(E_s E_{s-1} \dots E_2 E_1) \det(B) = \det(A) \det(B).
 \end{aligned}$$

10. Let  $A \in \mathbb{M}_n(\mathbb{R})$ . Then prove that  $\det(A) = \det(A^T)$ . If  $A \in \mathbb{M}_n(\mathbb{C})$  then  $\overline{\det(A)} = \det(A^*)$ .

**Solution:**

Recall that, a square matrix can be reduced to an upper form using elementary row operations. In other words, there exist elementary matrices  $E_s, E_{s-1}, \dots, E_2, E_1$  such that

$$E_s E_{s-1} \dots E_2 E_1 A = U.$$

where  $U$  is an upper triangular matrix and each  $E_k$ ,  $1 \leq k \leq s$ , is either an elementary matrix  $E_{ij}$  or  $E_{ij}(c)$  for some  $i, j$  and  $c$ . We therefore have

$$EA = U$$

with  $\det(E) = \det(E^T) = \pm 1$ . We also have  $\det(U) = \det(U^T)$  for the upper triangular matrix  $U$ . Thus

$$\det(A) = \frac{\det(U)}{\det(E)},$$

and

$$A^T E^T = U^T$$

yields

$$\det(A^T) = \frac{\det(U^T)}{\det(E^T)} = \frac{\det(U)}{\det(E)} = \det(A).$$

Thus,  $\det(\bar{A}) = \det((\bar{A})^T) = \det(A^*)$ . The equality  $\det(\bar{A}) = \overline{\det A}$  follows from the determinant formula:

$$\det(\bar{A}) = \sum_{\sigma \in S_n} (\text{sgn } \sigma) \overline{a_{1\sigma(1)}} \overline{a_{2\sigma(2)}} \dots \overline{a_{n\sigma(n)}} = \overline{\sum_{\sigma \in S_n} (\text{sgn } \sigma) a_{1\sigma(1)} a_{2\sigma(2)} \dots a_{n\sigma(n)}} = \overline{\det(A)}.$$

11. Let  $A$  be an  $n \times n$  matrix. Prove that

- (a) If  $A^2 = \mathbf{0}$  then  $A$  is not invertible (singular).

**Solution:** Let  $A$  be invertible. Then there exists  $B$  such that  $AB = I$ . Multiplying on the left by  $A$  gives  $A(AB) = A \cdot I = A$ . Thus,  $A = A^2B = \mathbf{0}B = \mathbf{0}$ , a contradiction, as the zero-matrix is not invertible.

**Alternate Solution:**  $0 = \det(A^2) = (\det A)^2 \Rightarrow \det A = 0$ .

- (b) If  $A^2 = A$ ,  $A \neq I$  then  $A$  is singular.

**Solution:** If  $A$  is nonsingular, then it is invertible. Then  $A = A^{-1}A^2 = A^{-1}A = I$ .

12. Can  $RREF([A|\mathbf{b}]) = \left[ \begin{array}{ccc|c} 1 & * & * & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{array} \right]$ ,  $\left[ \begin{array}{ccc|c} 0 & 1 & * & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{array} \right]$  or  $\left[ \begin{array}{ccc|c} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{array} \right]$ ? Explain.

Now, recall the matrices  $A_j$ 's, for  $1 \leq j \leq 3$  (defined to state the Cramer's rule for solving the linear system  $A\mathbf{x} = \mathbf{b}$ ), that are obtained by replacing the  $j$ -th column of  $A$  by  $\mathbf{b}$ . Then, we see that the above system has **NO solution** even though  $\det(A) = 0 = \det(A_j)$ , for  $1 \leq j \leq 3$ .

13. Let  $A$  be an  $n \times n$  matrix. Prove that the following statements are equivalent:

- (a)  $\det(A) \neq 0$ .
- (b)  $A$  is invertible.
- (c) The homogeneous system  $A\mathbf{x} = \mathbf{0}$  has only the trivial solution.
- (d) The row-reduced echelon form of  $A$  is  $I_n$ .
- (e)  $A$  is a product of elementary matrices.
- (f) The system  $A\mathbf{x} = \mathbf{b}$  has a unique solution for every  $\mathbf{b}$ .
- (g) The system  $A\mathbf{x} = \mathbf{b}$  is consistent for every  $\mathbf{b}$ .

**Solution:**

13a  $\implies$  13b

By, definition, whenever  $\det(A) \neq 0$ ,  $A^{-1} = \frac{C^T}{\det(A)}$ , where  $C$  is the co-factor matrix.

13b  $\implies$  13a

As  $A$  is invertible,  $AA^{-1} = I_n$  and hence  $\det(A)\det(A^{-1}) = \det(I_n) = 1$ . Hence,  $\det(A) \neq 0$ .

13b  $\implies$  13c

As  $A$  is invertible,  $A^{-1}A = I_n$ . Let  $\mathbf{x}_0$  be a solution of the homogeneous system  $A\mathbf{x} = \mathbf{0}$ . Then,

$$\mathbf{x}_0 = I_n\mathbf{x}_0 = (A^{-1}A)\mathbf{x}_0 = A^{-1}(A\mathbf{x}_0) = A^{-1}\mathbf{0} = \mathbf{0}.$$

Thus,  $\mathbf{0}$  is the only solution of the homogeneous system  $A\mathbf{x} = \mathbf{0}$ .

13c  $\implies$  13d

Let  $\mathbf{x}^T = [x_1, x_2, \dots, x_n]$ . As  $\mathbf{0}$  is the only solution of the linear system  $A\mathbf{x} = \mathbf{0}$ , the final equations are  $x_1 = 0, x_2 = 0, \dots, x_n = 0$ . These equations can be rewritten as

$$\begin{aligned} 1 \cdot x_1 + 0 \cdot x_2 + 0 \cdot x_3 + \dots + 0 \cdot x_n &= 0 \\ 0 \cdot x_1 + 1 \cdot x_2 + 0 \cdot x_3 + \dots + 0 \cdot x_n &= 0 \\ 0 \cdot x_1 + 0 \cdot x_2 + 1 \cdot x_3 + \dots + 0 \cdot x_n &= 0 \\ &\vdots \\ 0 \cdot x_1 + 0 \cdot x_2 + 0 \cdot x_3 + \dots + 1 \cdot x_n &= 0. \end{aligned}$$

That is, the final system of homogeneous system is given by  $I_n \cdot \mathbf{x} = \mathbf{0}$ . Or equivalently, the row-reduced echelon form of the augmented matrix  $[A \ \mathbf{0}]$  is  $[I_n \ \mathbf{0}]$ . That is, the row-reduced echelon form of  $A$  is  $I_n$ .

13d  $\implies$  13e

Suppose that the row-reduced echelon form of  $A$  is  $I_n$ . Then there exist elementary matrices  $E_1, E_2, \dots, E_k$  such that

$$E_1 E_2 \dots E_k A = I_n. \quad (1)$$

Now, the matrix  $E_j^{-1}$  is an elementary matrix and is the inverse of  $E_j$  for  $1 \leq j \leq k$ . Therefore, successively multiplying Equation (1) on the left by  $E_1^{-1}, E_2^{-1}, \dots, E_k^{-1}$ , we get

$$A = E_k^{-1} E_{k-1}^{-1} \dots E_2^{-1} E_1^{-1}$$

and thus  $A$  is a product of elementary matrices.

13e  $\implies$  13b

Suppose  $A = E_1 E_2 \dots E_k$ ; where the  $E_i$ 's are elementary matrices. As the elementary matrices are invertible and the product of invertible matrices is also invertible, we get the required result.

13b  $\implies$  13f

Observe that  $\mathbf{x}_0 = A^{-1}\mathbf{b}$  is the unique solution of the system  $A\mathbf{x} = \mathbf{b}$ .

13f  $\implies$  13g

The system  $A\mathbf{x} = \mathbf{b}$  has a solution and hence by definition, the system is consistent.

13g  $\implies$  13b

For  $1 \leq i \leq n$ , define  $\mathbf{e}_i = (0, \dots, 0, \underbrace{1}_{i^{\text{th}} \text{ position}}, 0, \dots, 0)^T$ , and consider the linear system

$A\mathbf{x} = \mathbf{e}_i$ . By assumption, this system has a solution, say  $\mathbf{x}_i$ , for each  $i$ ,  $1 \leq i \leq n$ . Define a matrix  $B = [\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n]$ . That is, the  $i^{\text{th}}$  column of  $B$  is the solution of the system  $A\mathbf{x} = \mathbf{e}_i$ . Then

$$AB = A[\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n] = [A\mathbf{x}_1, A\mathbf{x}_2, \dots, A\mathbf{x}_n] = [\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n] = I_n.$$

Therefore, the matrix  $A$  is invertible.

14.  $A \in \mathbb{M}_n(\mathbb{C})$ . Then  $\det(A) = 0$  if and only if the system  $A\mathbf{x} = \mathbf{0}$  has a non-trivial solution.

15. **(T)** Let  $A$  be an  $n \times n$  matrix. Then, the two statements given below cannot hold together.

- (a) The system  $A\mathbf{x} = \mathbf{b}$  has a solution for every  $\mathbf{b}$ .  
 (b) The system  $A\mathbf{x} = \mathbf{0}$  has a non-trivial solution.
16. Suppose the  $4 \times 4$  matrix  $M$  has 4 equal rows all containing  $a, b, c, d$ . We know that  $\det(M) = 0$ . The problem is to find by any method

$$\det(I + M) = \begin{vmatrix} 1+a & b & c & d \\ a & 1+b & c & d \\ a & b & 1+c & d \\ a & b & c & 1+d \end{vmatrix}.$$

**Solution:** Subtracting row 1 from rows 2, 3 and 4, we get

$$\det(I + M) = \begin{vmatrix} 1+a & b & c & d \\ -1 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ -1 & 0 & 0 & 1 \end{vmatrix}.$$

Now, adding columns 2, 3 and 4 to column 1, we get

$$\det(I + M) = \begin{vmatrix} 1+a+c+d & a & b & d \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{vmatrix}.$$

Thus,  $\det(I + M) = 1 + a + b + c + d$ .

17. The numbers 1375, 1287, 4191 and 5731 are all divisible by 11. Prove that 11 also divides the determinant of the matrix

$$\begin{bmatrix} 1 & 1 & 4 & 5 \\ 3 & 2 & 1 & 7 \\ 7 & 8 & 9 & 3 \\ 5 & 7 & 1 & 1 \end{bmatrix}.$$

**Solution:** Adding  $1000 \times \text{Row}_1$ ,  $100 \times \text{Row}_2$ ,  $10 \times \text{Row}_3$  to  $\text{Row}_4$ , we have

$$\begin{vmatrix} 1 & 1 & 4 & 5 \\ 3 & 2 & 1 & 7 \\ 7 & 8 & 9 & 3 \\ 5 & 7 & 1 & 1 \end{vmatrix} = \begin{vmatrix} 1 & 1 & 4 & 5 \\ 3 & 2 & 1 & 7 \\ 7 & 8 & 9 & 3 \\ 1375 & 1287 & 4191 & 5731 \end{vmatrix}.$$

Since  $\text{Row}_4$  is divisible by 11, the determinant is divisible by 11.

18. Compute determinant of  $\begin{bmatrix} 1 & x_1 & x_1^2 & x_1^3 & x_1^4 \\ 1 & x_2 & x_2^2 & x_2^3 & x_2^4 \\ 1 & x_3 & x_3^2 & x_3^3 & x_3^4 \\ 1 & x_4 & x_4^2 & x_4^3 & x_4^4 \\ 1 & x_5 & x_5^2 & x_5^3 & x_5^4 \end{bmatrix}$ .

**Solution:** We give the solution for the general case. Let

$$A_n = \begin{bmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{n-1} \\ \cdot & \cdot & \cdot & \cdots & \cdot \\ \cdot & \cdot & \cdot & \cdots & \cdot \\ 1 & x_n & x_n^2 & \cdots & x_n^{n-1} \end{bmatrix}.$$

If  $n = 2$ ,  $\det(A_2) = x_2 - x_1$ . We will prove that

$$\det(A_n) = \prod_{i < j} (x_j - x_i).$$

Assume the result for  $n - 1$  and define

$$F(x) = \begin{vmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{n-1} \\ \cdot & \cdot & \cdot & \cdots & \cdot \\ \cdot & \cdot & \cdot & \cdots & \cdot \\ 1 & x & x^2 & \cdots & x^{n-1} \end{vmatrix}.$$

Then  $F$  is a polynomial of degree  $n - 1$  with roots  $x_1, x_2, \dots, x_{n-1}$ . So,  $F(x) = c \prod_{i=1}^{n-1} (x - x_i)$  where  $c$  is coefficient of  $x^{n-1}$  which is clearly  $\det(A_{n-1})$ . Therefore,

$$F(x) = \det(A_{n-1}) \prod_{i=1}^{n-1} (x - x_i).$$

The result follows for  $n$  as

$$\det(A_n) = F(x_n) = \det(A_{n-1}) \prod_{i=1}^{n-1} (x_n - x_i).$$